

Week 3 Report: Hypothesis Testing

Data Analytics & Machine Learning Research Lab

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23rd January 2025

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1 Hypothesis testing \bar{u}

1.1 Problem

- Given two sets of values, X and Y , each containing 50 elements, perform a hypothesis testing to determine whether their means are equal.
- The **Null hypothesis** (H_0) is:

$$\mu_x = \mu_y$$

- The **Alternative Hypothesis** (H_1) is:

$$\mu_x \neq \mu_y$$

1.2 Analysis

- Because we compare 2 sets of values with 50 elements each, we will use Z-Test. Specifically, we compare the mean value of 2 separate sets so we use the 2-Sample Z-Test:

$$z = \frac{\bar{x} - \bar{y}}{\sqrt{\frac{\sigma_x^2}{n_x} + \frac{\sigma_y^2}{n_y}}}$$

\bar{x}, \bar{y} : Mean value of the two sets

σ_x^2, σ_y^2 : Variance of the two sets

n_x, n_y : Size of the two sets

- Calculate variance:

$$\sigma^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$$

- Distribution of Z value: **Gaussian Distribution**

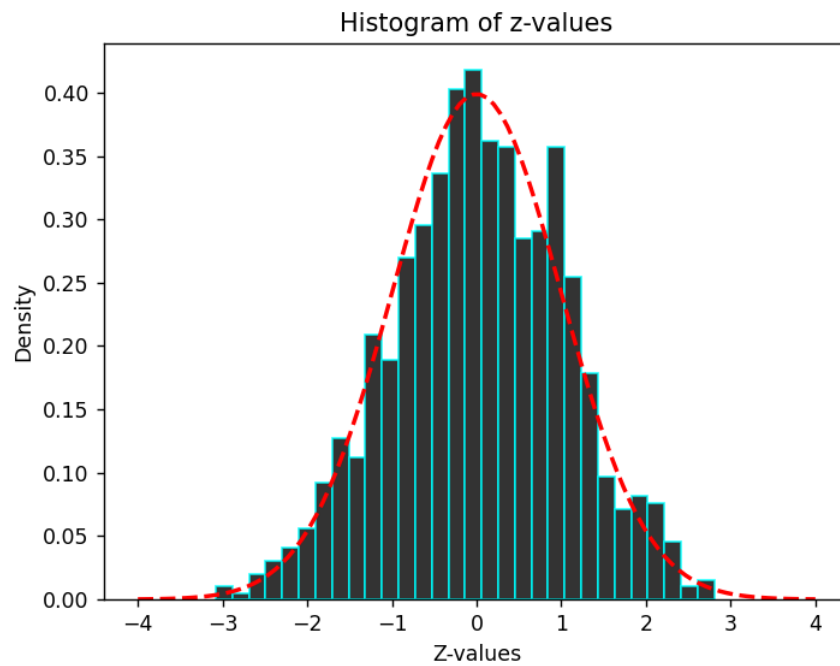


Figure 1: Z Distribution

- P_value graph:

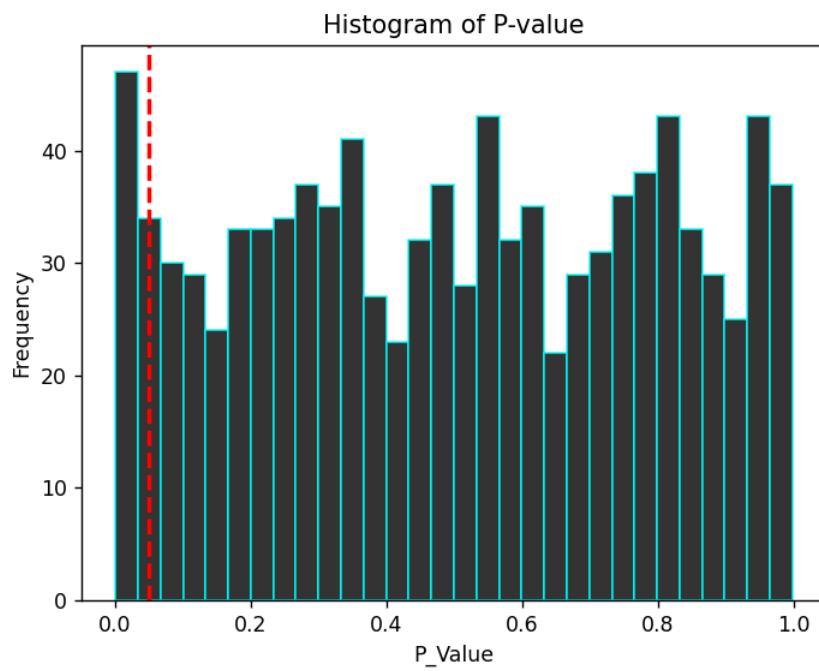


Figure 2: P_value

1.3 Coding

2 Hypothesis testing Linear Regression

2.1 Problem

- Given a linear regression model, use hypothesis testing to determine whether the coefficients in a model are different from 0 (**Important**) or equal 0 (**Unnecessary**)

- The **Null hypothesis** (H_0) is:

$$\beta_i = 0$$

- The **Alternative hypothesis** (H_1) is:

$$\beta_i \neq 0$$

2.2 Analysis

- We use Z-Test to calculate p-value

$$z_i = \frac{\hat{\beta}_i}{SE(\hat{\beta}_i)}$$

- Calculate **Residuals**: $\hat{\beta}_i = \hat{\omega}_i - 0$

- Calculate **Standard Error**: $SE(\hat{\beta}_i) = \sqrt{Cov(\hat{\beta}_i)}$

Covariance Matrix:

$$Cov(\hat{\beta}) = \sigma^2 (X^T X)^{\dagger}$$

Variance:

$$\frac{RSS}{n - k}$$

n: number of elements

k: number of features + bias

RSS (Residuals Sum of Square):

$$\sum_{i=1}^n (y_i - \hat{y}_i)^2$$

- Calculate **P-value**:

$$P_value = 2 * \min [1 - \Phi(|z_i|), \Phi(|z_i|)]$$

$\Phi(x)$: CDF

- **Coefficients**: [1, 2, 0, 0.8, 0, 1.5]
- **Distribution**: Gaussian Distribution

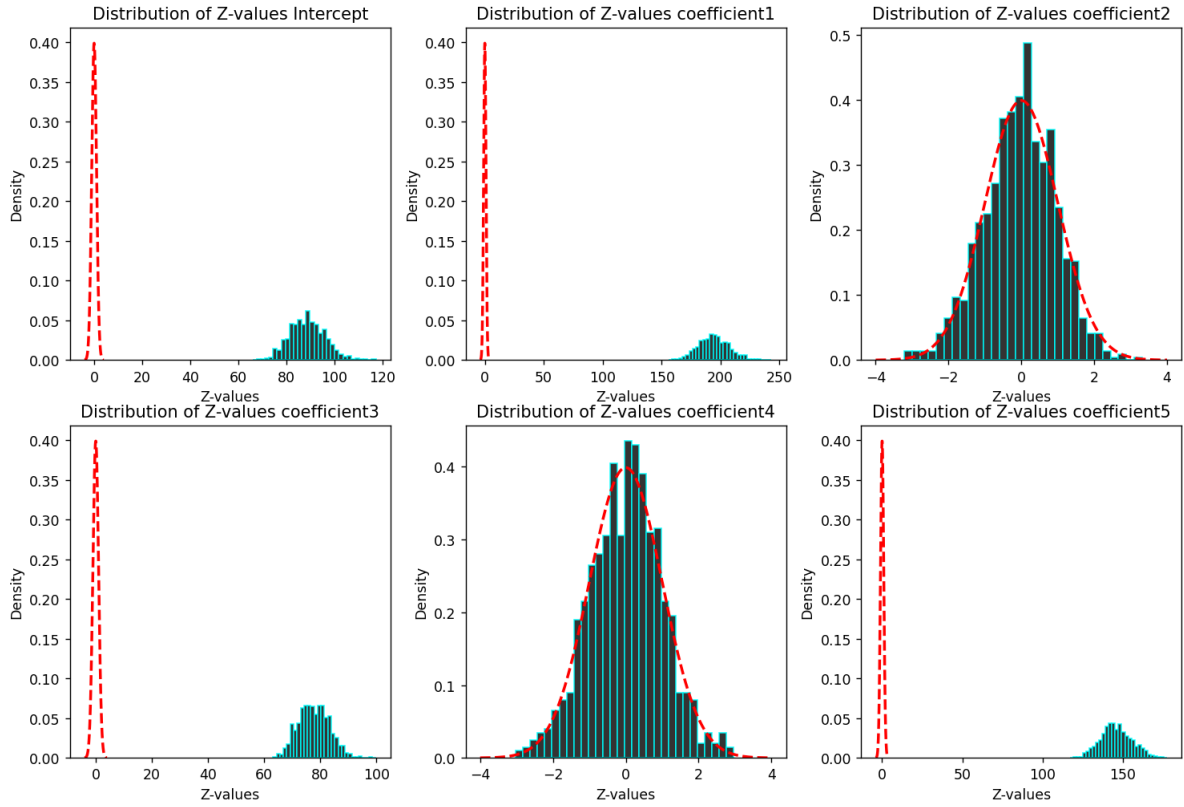


Figure 3: Linear Regression Z-Value

- P_value Graph

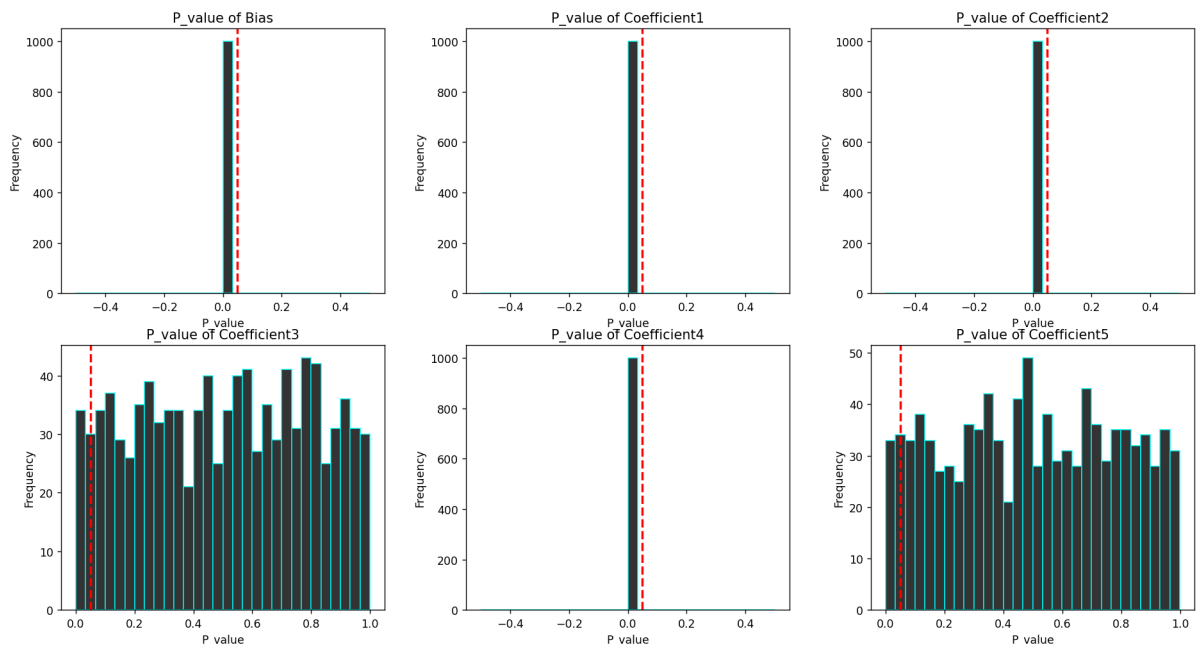


Figure 4: Linear Regression P_value

2.3 Code